The Power and Limitations of Item Price Combinatorial Auctions

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Talk Structure

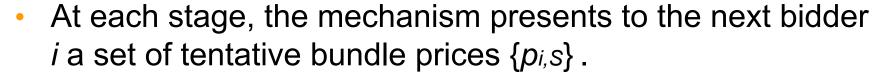
- Types of Iterative Auctions
- Computational Power of Item-Price Auctions
- Complexity: item-price vs. Bundle-price auctions
- Some approximations by item-price auctions

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Combinatorial Auctions

- m indivisible non-identical items for sale
- n bidders compete for subsets of these items
- Each bidder i has a valuation for each set of items:
 vi(S) = value that i assigns to acquiring the set S
 - vi is non-decreasing ("free disposal")
 - $Vi(\emptyset) = 0$
- **Objective:** Find a partition $(S_1...S_n)$ of $\{1...m\}$ that maximizes the social welfare: $\sum_i v_i(S_i)$
- Issues: communication, allocation, strategies

Iterative Auction Mechanisms



- The bidder at this stage should bid for the set S that maximizes his utility at these prices: vi(S) - pi,s.
- The mechanism rules determine: the prices at each stage, who bids at each stage, when to stop. Upon termination it should determine the final allocation and the payments.
- Types of natural restrictions:
 - Item-prices (linear prices): $p_{i,s} = \sum_{j \in S} p_{i,j}$
 - Anonymous prices: pi,s = pi',s
 - Ascending Auctions: pi,s non-decreasing with time

Why "should" bidders follow protocol?

- Many possible answers, with varying degrees of plausibility:
 - Incentive Compatible (in dominant strategies)
 - Charge VCG prices ex-post-Nash incentive compatibility
 - Proxies whether actual or cryptographically simulated
 - Myopic this is exactly why these bids are intuitive
 - Obedient A purely computational perspective
- In this talk we stay agnostic about incentives
 - Present whatever incentive properties obtained in each case
 - Impossibility results apply regardless of incentives

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Can the efficient outcome be obtained?

- Parkes-Ungar and Ausubel-Milgron mechanisms reach the efficient allocation. They are ascending, nonanonymous, and use (non-linear) bundle-prices.
- Open: are there ascending anonymous mechanisms that reach the efficient allocation?
- Our question: are there item-price auctions that reach the efficient allocation?
 - If valuations are substitutes → yes

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- What about the general case?
- Walrasian equilibrium with item prices does not exist.

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Demand oracle view of item-price auctions

Demand Oracle Query:

- Input: m item prices: p1...pm
- Output: $D(p_1...p_m)$ -- Demand at these prices. I.e. the set S that maximizes v(S)- $\sum_{j \in S} p_j$
- A general item-price iterative auction may be viewed as an allocation algorithm whose input is a demand oracle for each bidder.

Lemma: A demand oracle can simulate a valuation oracle. (And, the simulation is computationally efficient.)

Valuation Oracle Query:

- Input: subset S
- Output: v(S)

Corollary: (Weird) item-price auctions can reach the efficient outcome and produce VCG prices.

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Simulating a Valuation Oracle

Algorithm for computing v(S) using a demand oracle:

marginal-valuation(j,S):

For all goods $j \in S$ set $p_j = 0$; for all other goods set $p_j = \infty$ Perform a binary search on p_j to find lowest value with $D(p_1...p_m) = S$

<u>v(S):</u>

```
Initialize: result=0
For all goods j \in S do
result \leftarrow result + marginal-valuation(j, \{j' \in S \mid j' < j\})
```

	v(a)	v(b)	v(ab)
Player 1	∈ (0,1)	∈ (0,1)	2
Player 2	2	2	2

- Finding an efficient allocation requires answering: v1(a) <> v1(b)?
- Assume wlog that p_{1b} rose to 1 before p_{1a} rose to 1.
- Until this time, no information was gained (answer always {ab}).
- From this time on, no information on v₁(b) can be gained.

Analyzing complexity of auction mechanisms

- Consider only informational costs; ignore computation
 - "preference elicitation", "communication complexity"
- Basic lower bound: every combinatorial auction requires exponential communication in worst case

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- How to compare mechanisms?
 - How well they perform in real applications
 - We don't know. Not enough data. Life is hard.
 - How well they theoretically perform on classes of valuations
 - Semantic classes: "substitutes", "sub-modular", ...
 - Syntactic classes: XORs of few bundles, ORs of few bundles, ...
- What can we measure:
 - How much information transfer is needed to find socially efficient outcome.
 - How close to optimal can we get using "reasonable" information transfer.

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Item-price vs. Bundle price auctions

- Bundle-price auctions can be exponentially faster then item-price auctions.
 - Assume all valuations are XOR bids of at most s bundles
 - No bidder will ever bid another bundle
 - All reasonable bundle-price ascending auctions will terminate in at most s*n*max-bid/min-bid-increment steps.
 - Theorem (Blum, Jackson, Sandholm, Zinkevich): exponentially many demandoracle queries are needed, when $s=\sqrt{m}$.
- Can item-price auctions be faster than bundle-price auctions?
- Depends what you count as "information" in bundle-price auctions
 - Allow concise representation of bundle prices → at least as powerful as item-price auctions
 - Require to list each non-0 bundle price → yes.

Finding a "hidden" subset

- Consider the following case:
 - V1(S) = 2|S|,
 - Except for a single "hidden" set H with $v_1(H) = 2|H| + 1$.
 - $v_2(S) = (1+\varepsilon)2|S|$, for all S.
- Efficient allocation requires finding H and giving it to 1.
- A Walrasian equilibrium with a price of (2+ε) per item exists, and can be found quickly by item-price auctions.
- Assume even |H|=m/2 is known by a bundle price auction.
- No information about the identity of H can be obtained unless p₁(S) ≥ |S| for all sets S of size m/2+1.
- But this requires exponentially many bundle prices.

Suggestions so far

- A hybrid approach may be better than either item-price auctions or pure bundle-price auctions.
- Allowing bundle-price auctions to represent bundle prices succinctly suffices.
- How succinctly?
 - Either item prices or bundle prices
 - Arbitrary ORs of sub-bundle prices
 - General formula in some bidding language
 - Most extreme case: the aggregated bid of all others so far
- Try to evaluate auction mechanisms according to what types of valuations they can handle with reasonable information transfer.
- Challenge: a mechanism that can handle any XOR/ORformula bids (in time polynomial in the bid size).

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Approximation Algorithm

For $c = \varepsilon$, 2ε , ..., max-bid

For each bidder iLet Si be the demand (if $not \phi$) where all item prices are cIf demand became ϕ for the first time allocate Si to i (taking away any items that were previously allocated)

If (total value of solution found) $< vi(\{1...m\})$ for some i then Ignore current solution and allocate everything to i

Theorem: This gives a min(n,O(\sqrt{m})) approximation **Theorem:** Any auction that gives a better approximation requires an exponential number of queries.

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Sub-modular Valuations

Algorithm:

Initialize, $pi,j = \infty$ for all bidders i and items jFor every item j do

Perform Dutch auction on item j; let i be the winner $Si \leftarrow Si \cup j$; pi,j = 0

Theorem: If all valuations are sub-modular then this is a 2-approximation.

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Theorem: Any auction requires an exponential amount of communication to find an exact solution even for submodular valuations.

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Auctions with Duplicate Items

- Assume that there are k units of each good.
- Each bidder wants at most one from each good type i.e.
 vi() is still a functions of sets (rather than multi-sets).

Online Algorithm:

```
Initialize item prices: p1=...=p_m = vmin / (2km)
For all bidders i in order of arrival:
Si \leftarrow Di(p1...pm)
For all items j \in Si do
pj = \rho pj (for some well chosen \rho)
```

Theorem: This auction is valid, incentive compatible, and gives as good approximation as computationally possible.